



GEOSPATIAL TECHNOLOGIES IN SUSTAINABLE WATER MANAGEMENT IN AGRICULTURE

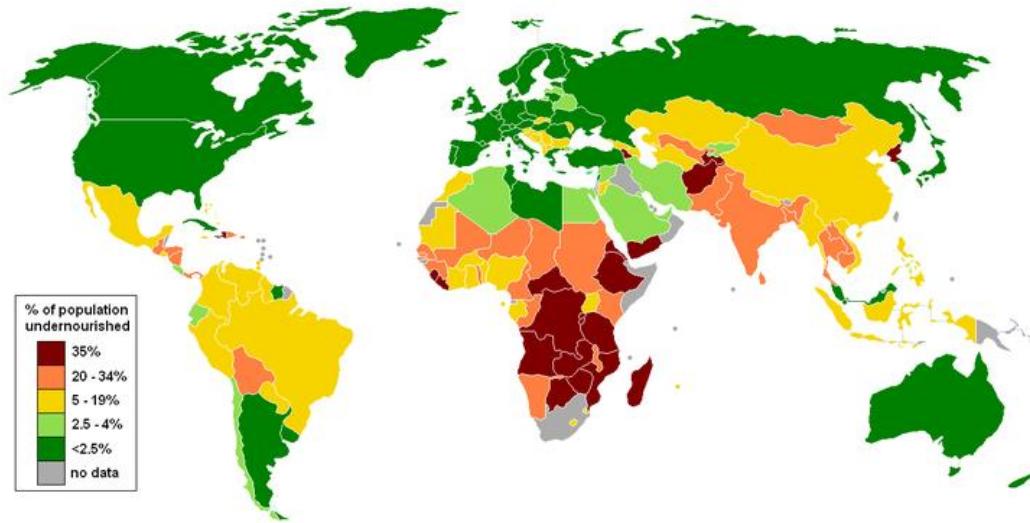
Elizaveta Khazieva



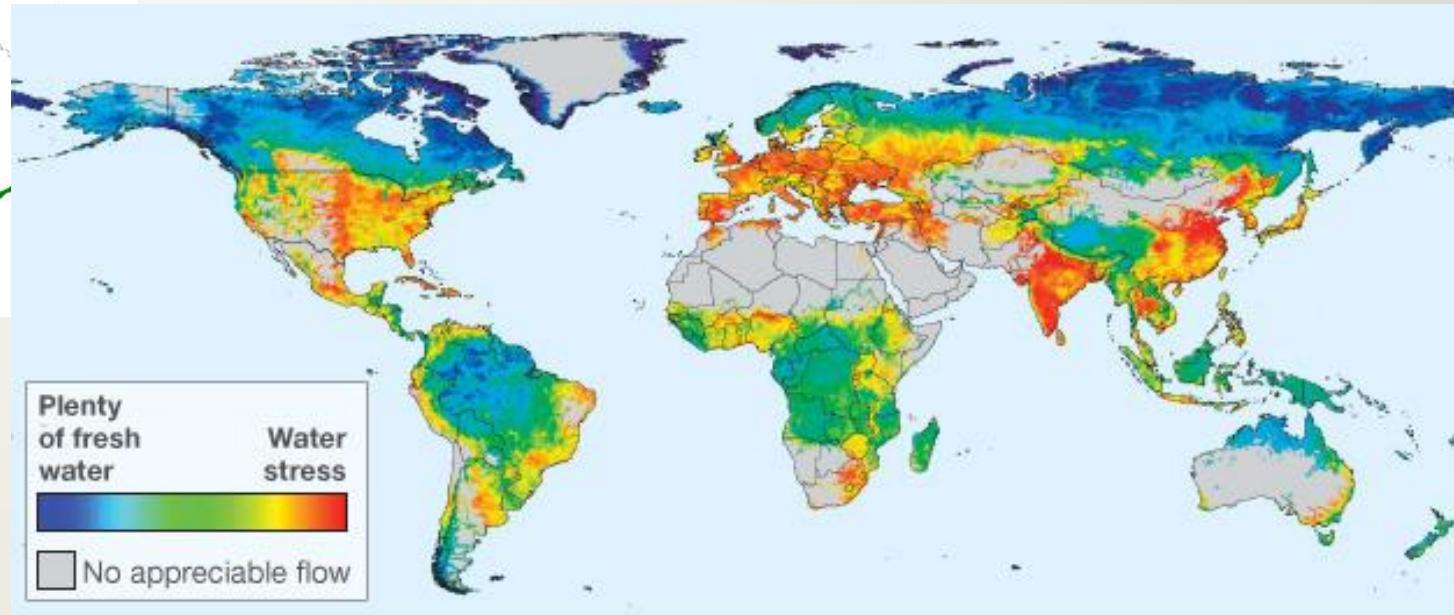
CENTRAL
EUROPEAN
UNIVERSITY

Agriculture and Food security

Map depicting the percentage of global undernourishment, 2008

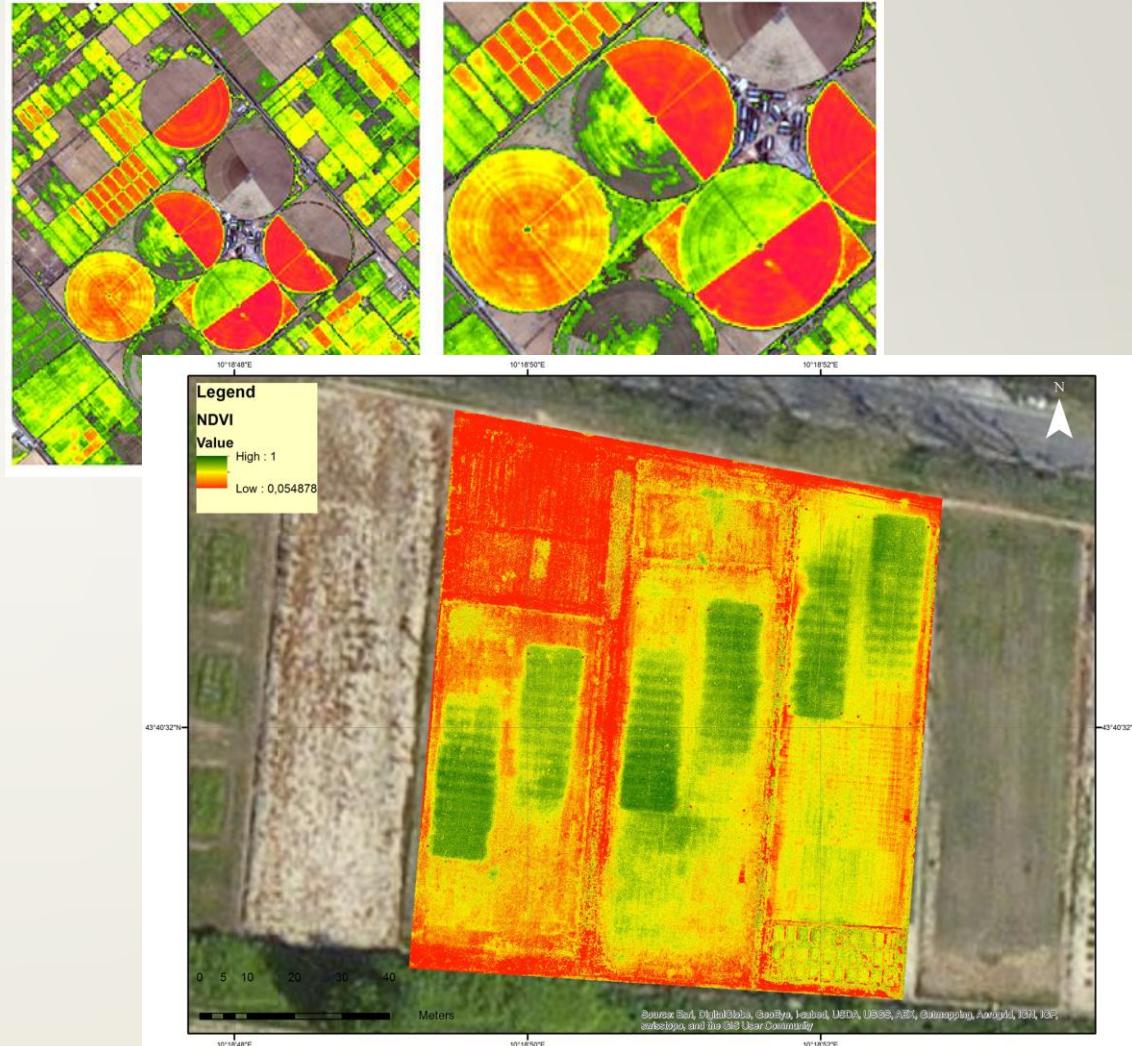


Global Water supply

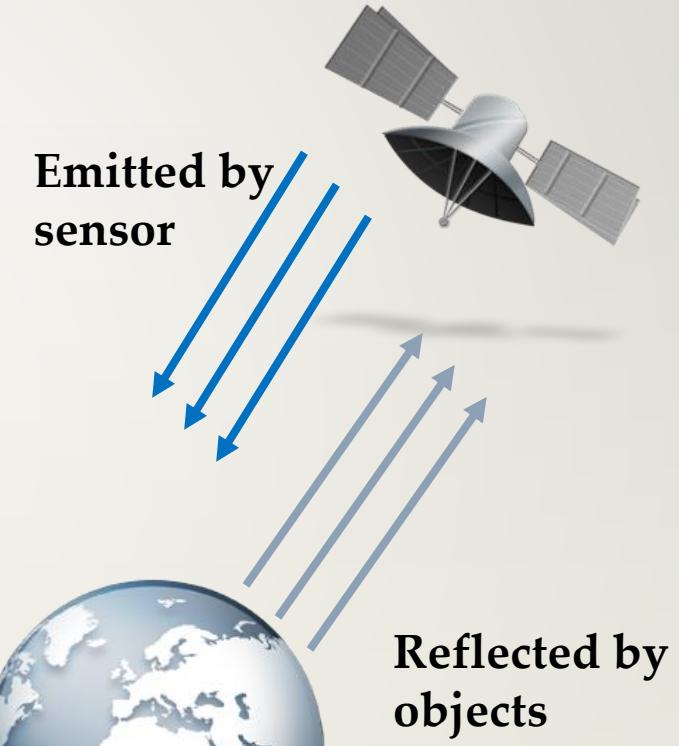
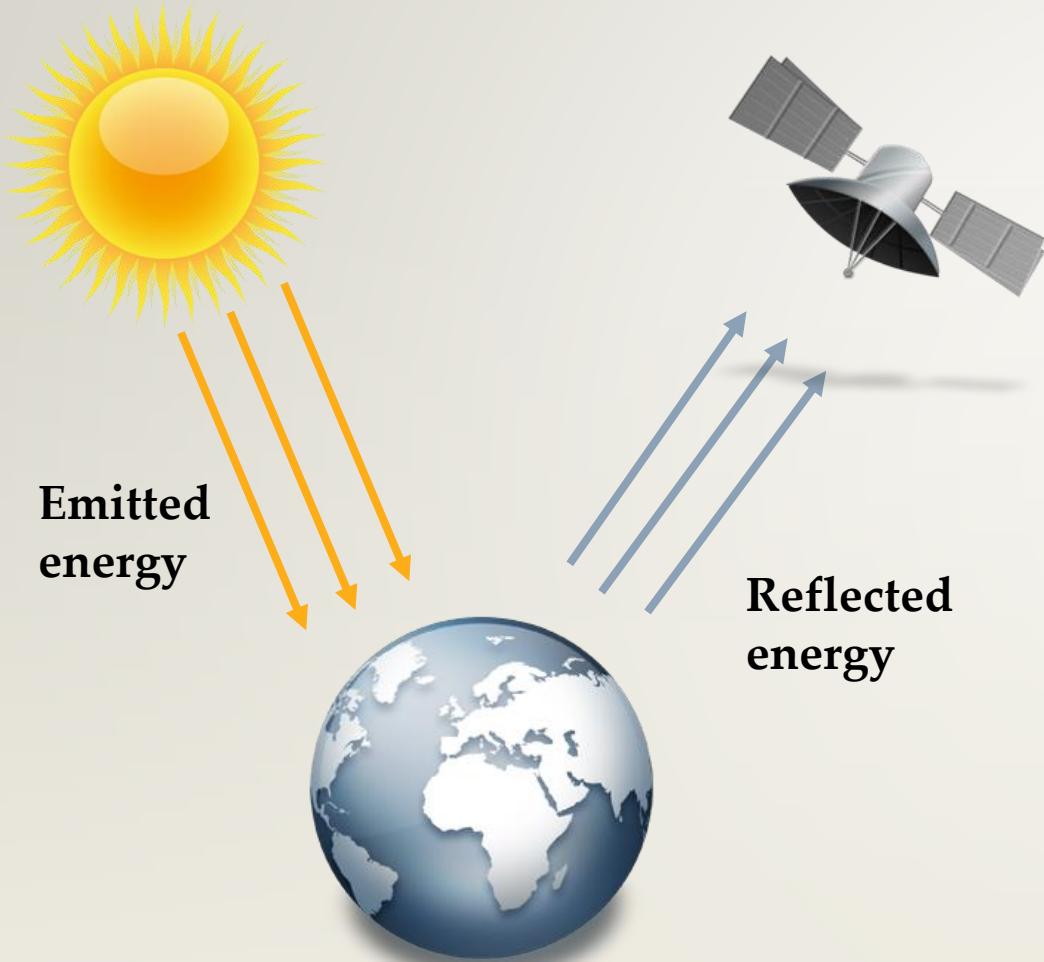


Geospatial technologies for agricultural management

- Crop identification
- Crop condition assessment
 - Early detection, diagnosis of plant diseases
 - Nutrient deficiency diagnostics
 - Water stress assessment
- Soil moisture estimation
- Irrigation planning
- Crop yield estimation
- Crop yield forecasting
- Precision agriculture



Passive vs active remote sensing



Potential data sources

<i>Data (active)</i>	<i>Description</i>	<i>Source</i>
ALOS PALSAR	5 modes (spatial resolution - from 7 m to 100 m.)	http://www.eorc.jaxa.jp/ALOS/en/gallery/new_arr.htm
Cosmo-SkyMed 1, 2, 3, 4	5 modes (spatial resolution - from 1 m to 100 m.)	http://www.e-geos.it/cosmo-skymed.html
Sentinel-1A	4 modes (spatial resolution - from 4 m to 80 m.)	https://scihub.copernicus.eu/dhus/#/home
Radarsat – 2	18 modes (spatial resolution - from 1 m to 100 m.)	http://www.asc-csa.gc.ca/eng/satellites/radarsat/radarsat-tableau.asp
Envisat	3 modes (spatial resolution - from 30 m to 1000 m.)	https://earth.esa.int/web/guest/missions/esa-operational-eo-missions/ers
TerraSAR-x	7 modes (spatial resolution - from 0.25 m to 40 m.)	http://terrasar-x-archive.infoterra.de/
Pleiades	4 spectral bands (spatial resolution - from 0.5 m to 2 m)	http://www.intelligence-airbusds.com/pleiades/

Passive (optical) sensors

Spectral bands combinations

Natural color image vs false color image



Spectral bands combinations



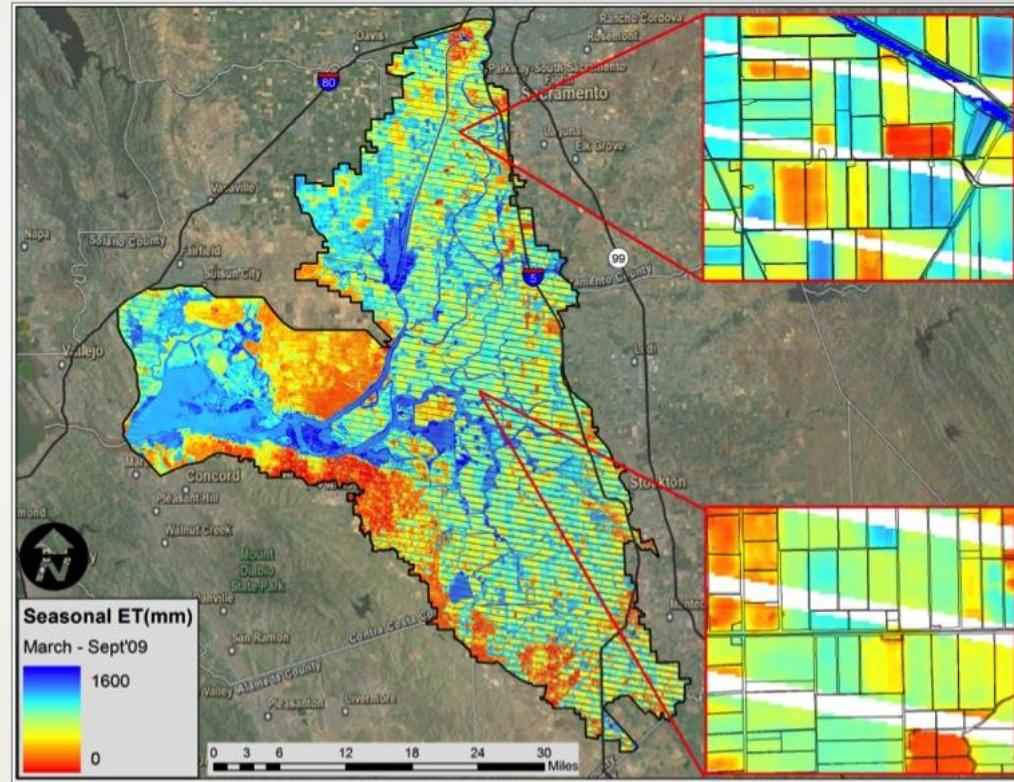
Landsat band combinations

False Colour 6,5,2
Vegetation

False Colour 7,6,4
Urban

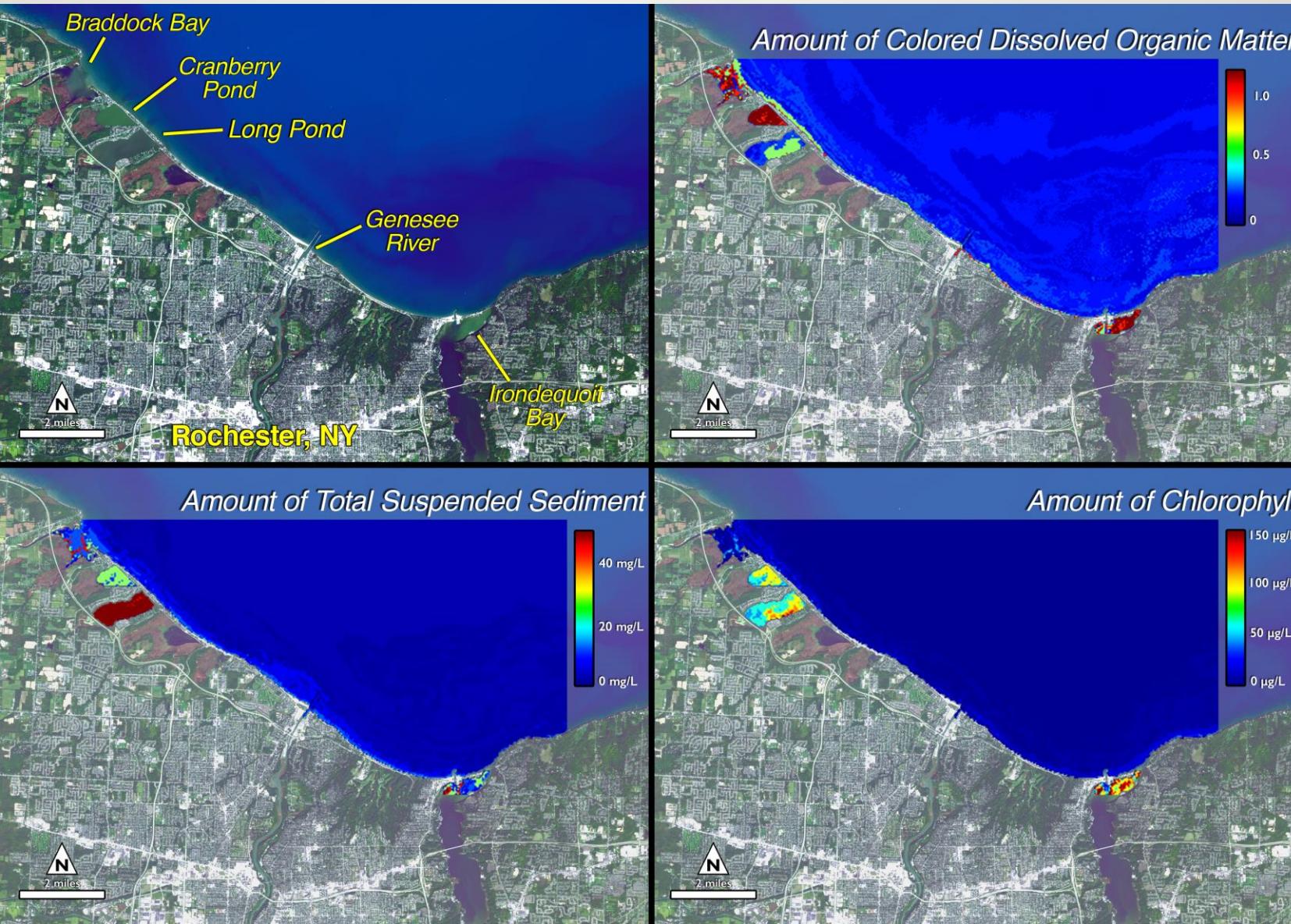
Colour IR 5,4,3
Vegetation

False Colour 5,6,4
Land/Water



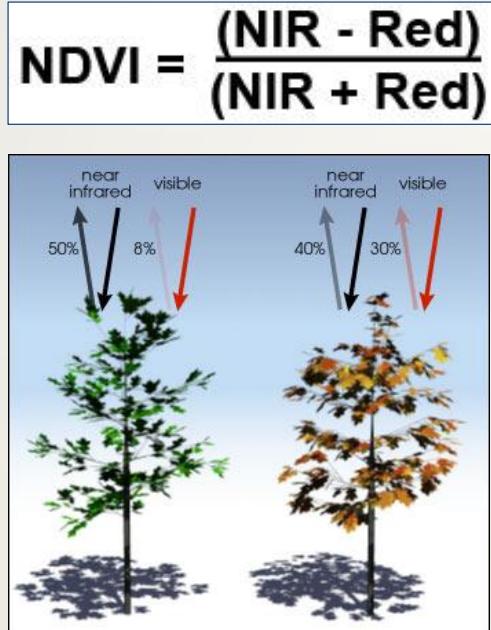
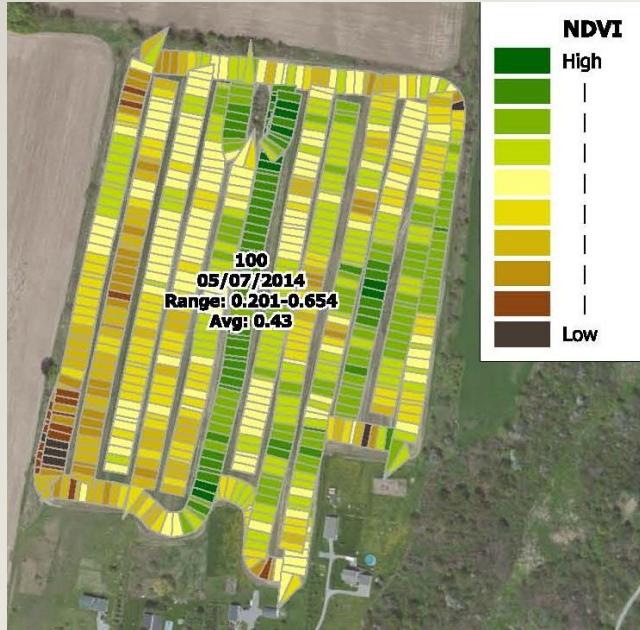
Seasonal evapotranspiration
crop growing season

Coastal water quality



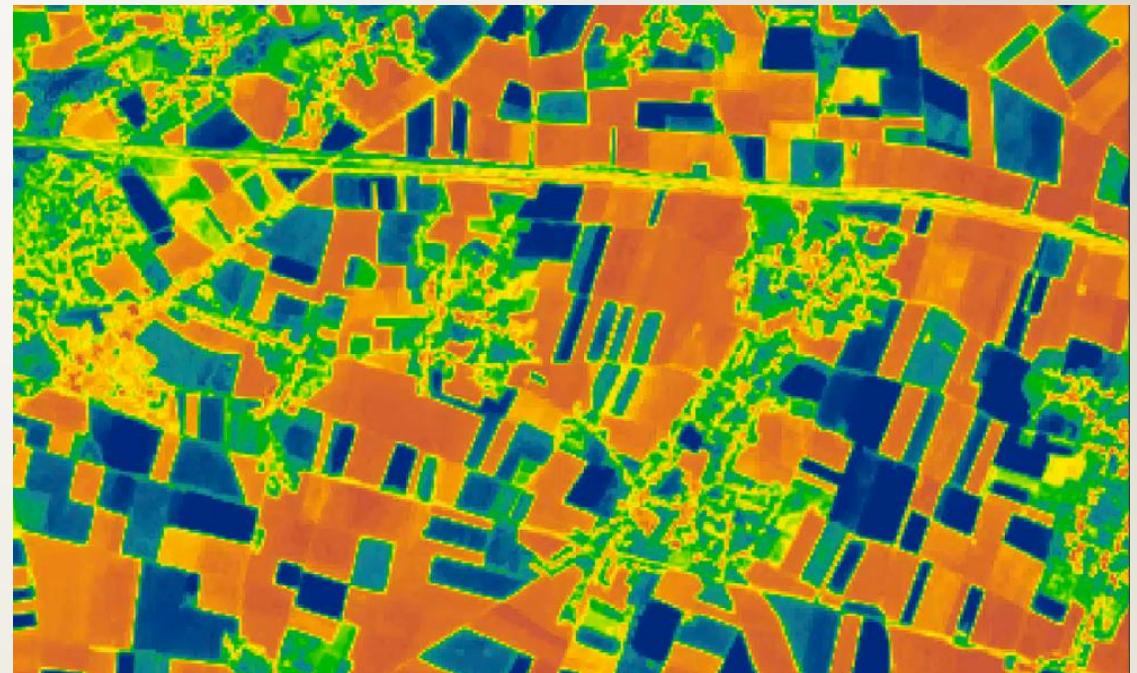
Indices for vegetation health assessment

NDVI
(Normalized Difference Vegetation Index)



NDWI
(Normalized Difference Water Index)

$$NDWI_t = \frac{NIR_t - SWIR_t}{NIR_t + SWIR_t}$$

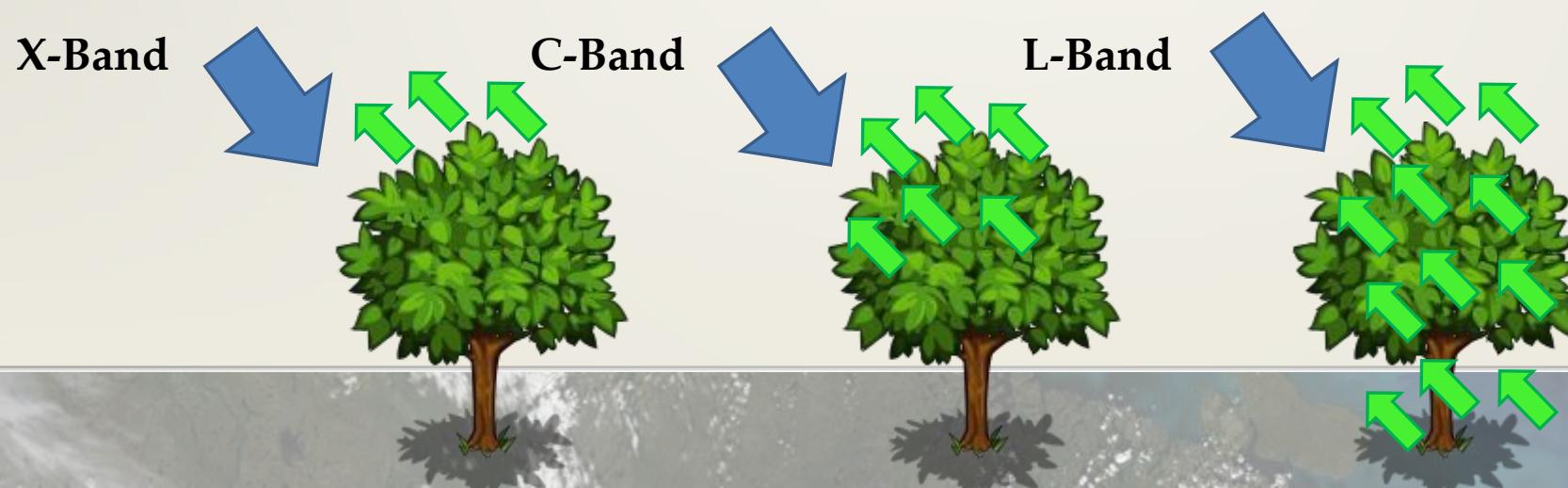


NDWI evolution from May to September

Active sensors

Commonly Used Frequency Bands

Frequency band	Frequency range	Application Example
P-Band	300 Mhz – 1 GHz	biomass, soil moisture, penetration
L-Band	1 GHz – 2 GHz	agriculture, forestry, soil moisture
C-Band	4 GHz – 8 GHz	ocean, agriculture
X-Band	8 GHz – 12 GHz	agriculture, ocean, high resolution radar
Ku-Band	14 GHz – 18 GHz	glaciology (snow cover mapping)
Ka-Band	27 GHz – 47 GHz	high resolution radars

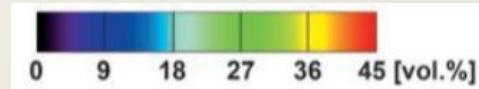


Radar images



Image in C-band

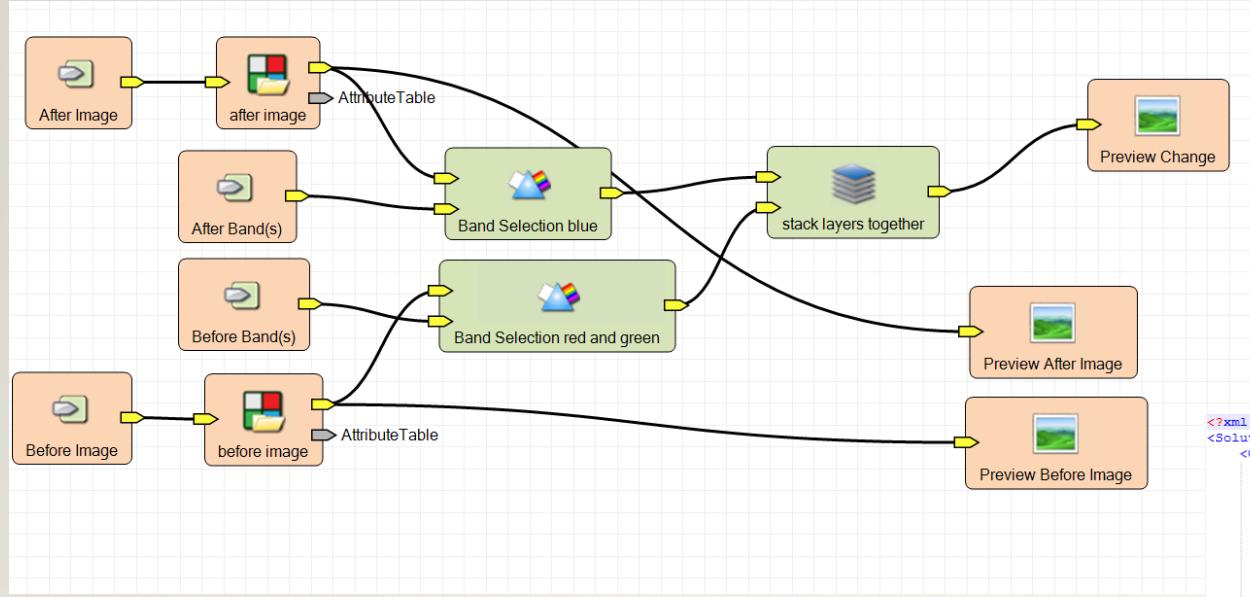
Image in L-band



soil moisture

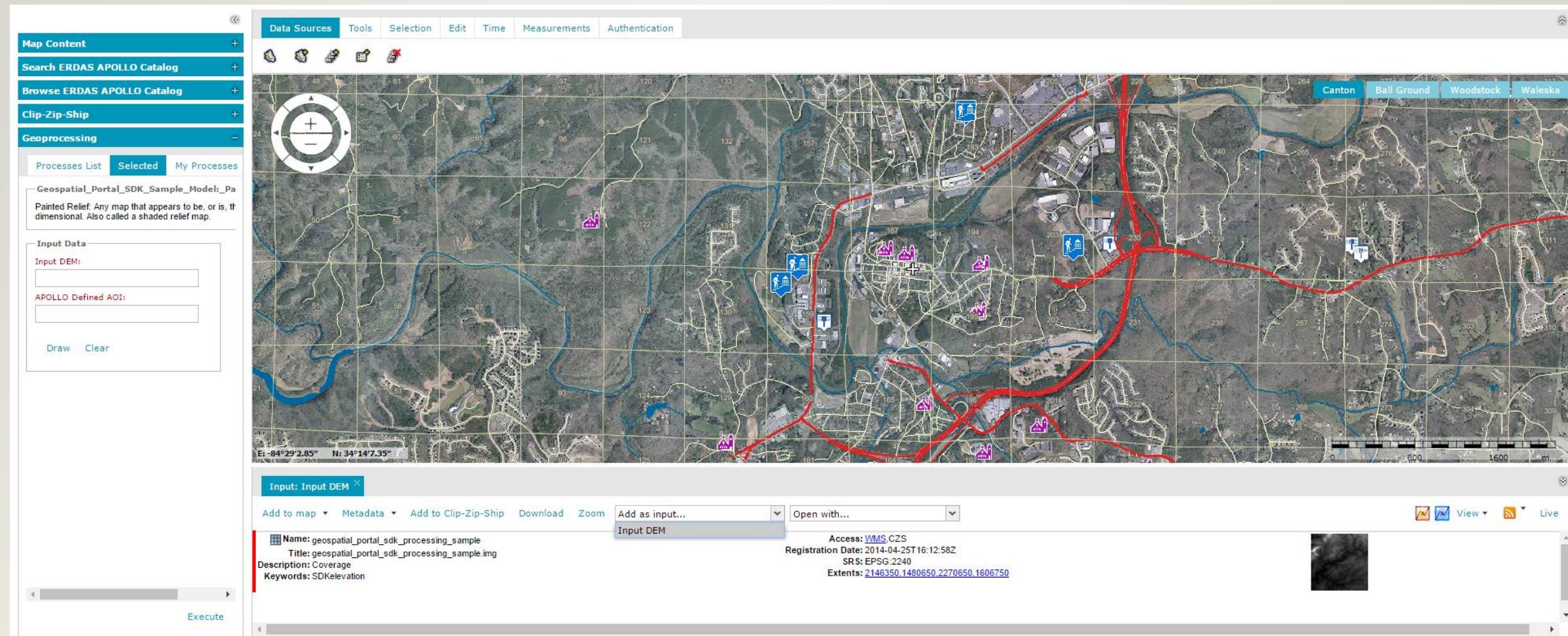
Data analysis and visualization

Spatial Modelling

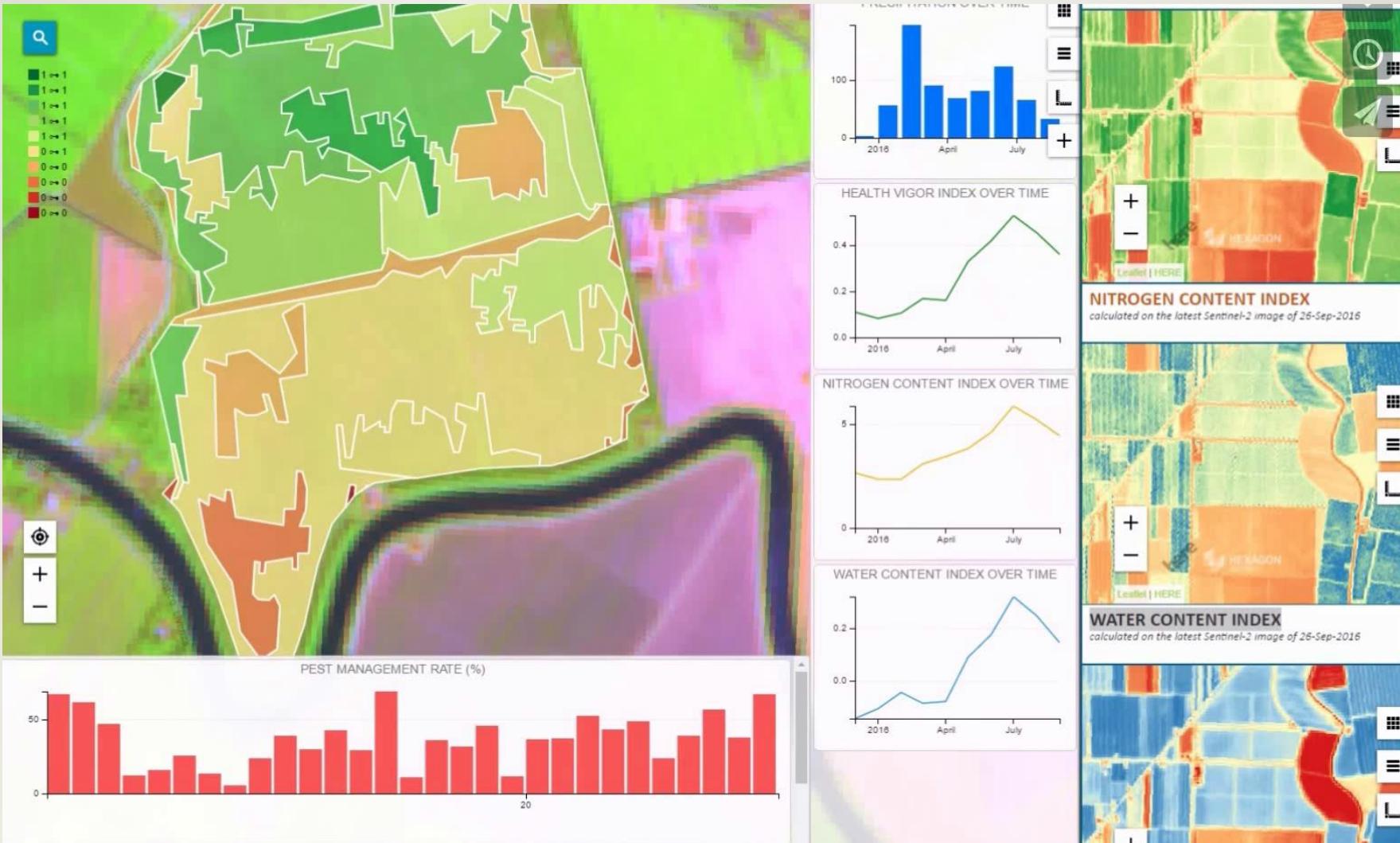


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<?xml version="1.0" encoding="UTF-8"?>
<Solution xmlns="http://tempuri.org/XMLSchema.xsd">
  <Operator Namespace="IMAGINE" Name="Process">
    <DisplayName>NDVI Categorised</DisplayName>
    <Description>Produces a six class dataset based on NDVI values from an input with NIR and Red bands</Description>
    <Port Name="Port Input 2">
      <DisplayName>Output File</DisplayName>
      <Description>Select file name of output categorised NDVI file</Description>
      <Input>true</Input>
      <Output>false</Output>
      <Hidden>false</Hidden>
      <Optional>false</Optional>
      <Data>
        <StringValue>c:/data/agro/agroosnova/images/landsat_vs_rapideye/june_2015_rapideye_ndvi_re_categor.img</StringValue>
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      <Attribute Name="AdditionalTypes">
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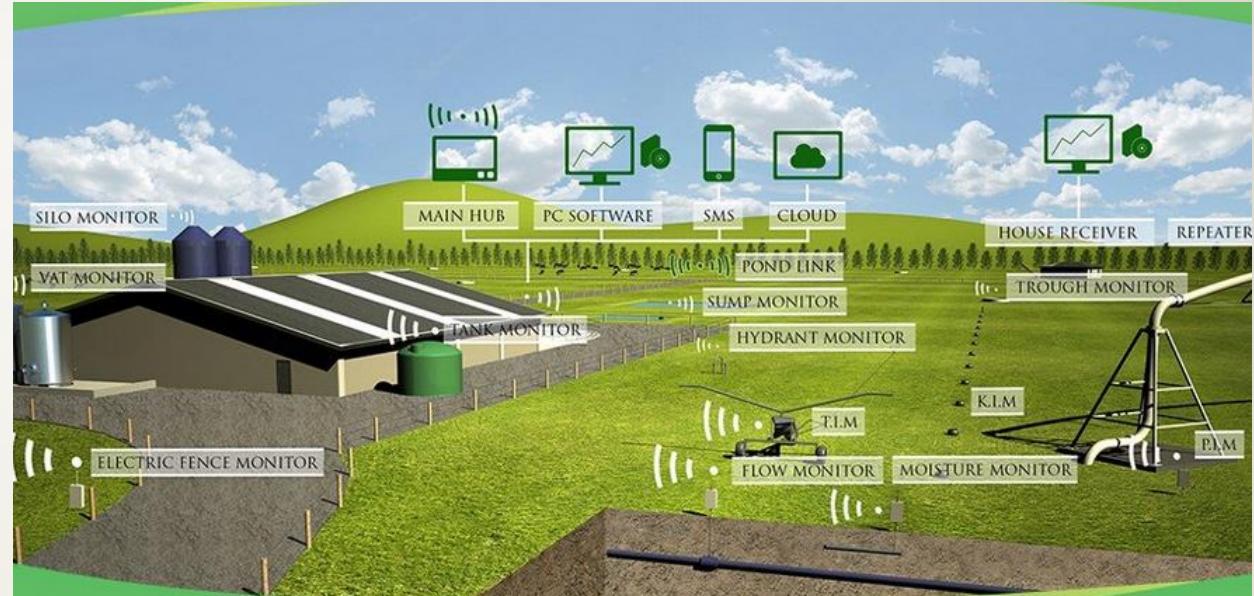
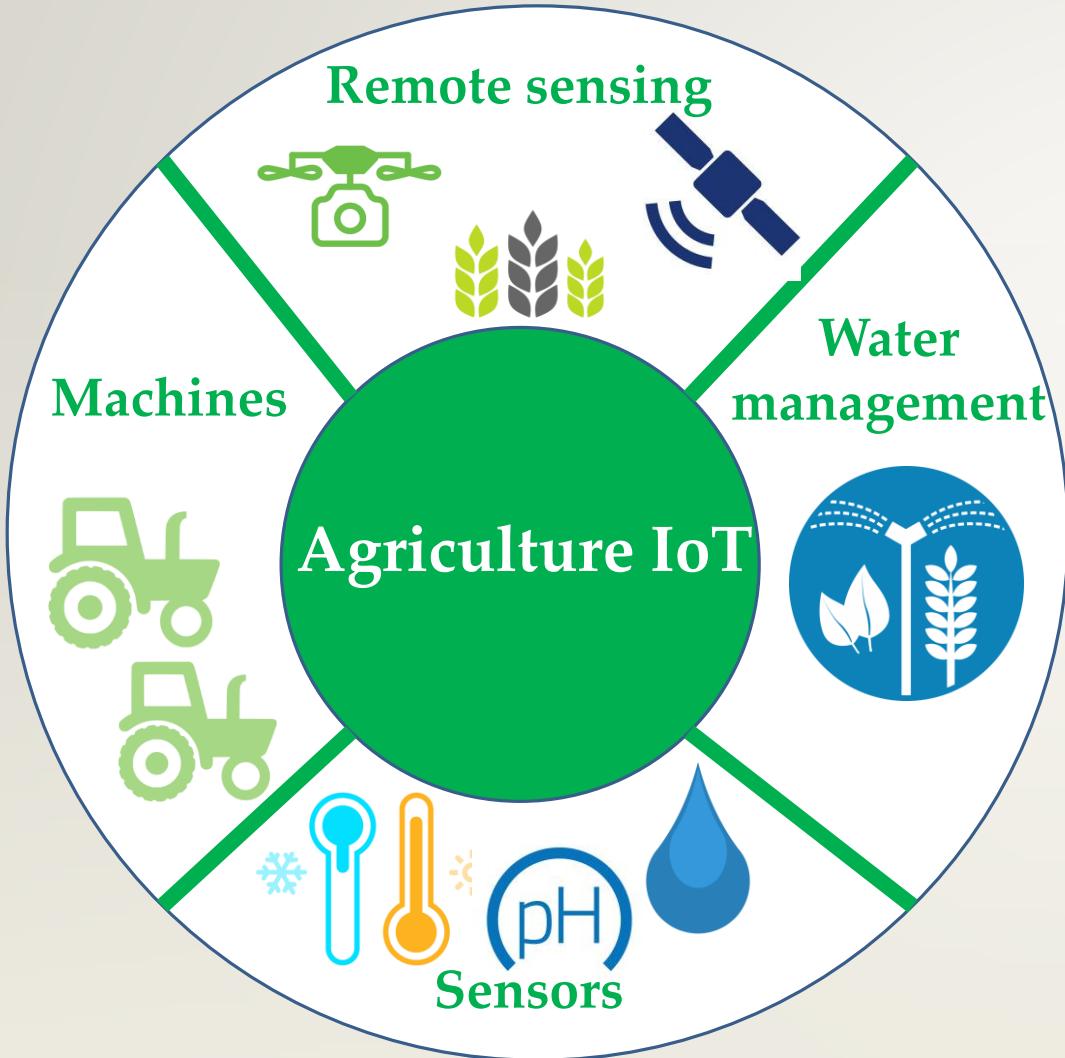
Spatial Modelling



Analysis Report



Internet of Thing in agriculture



Internet of Thing in agriculture



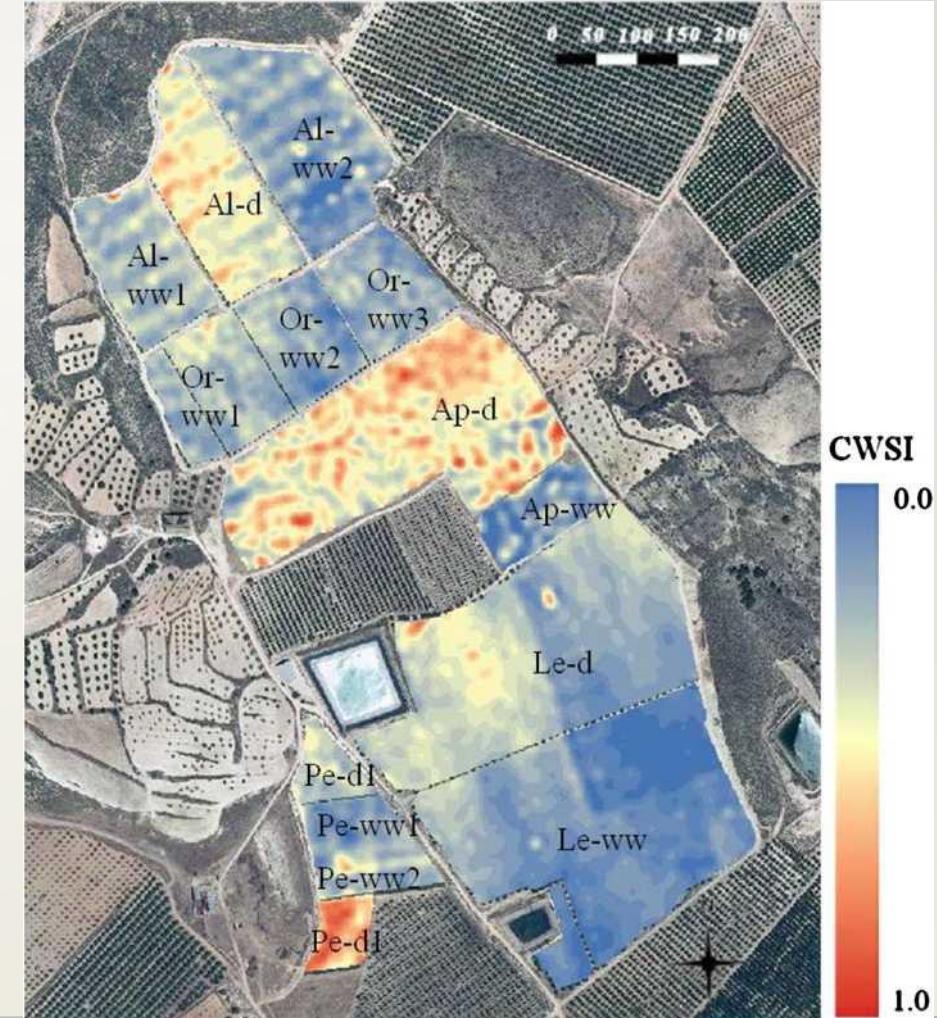
Projects and case studies

Using thermal imagery

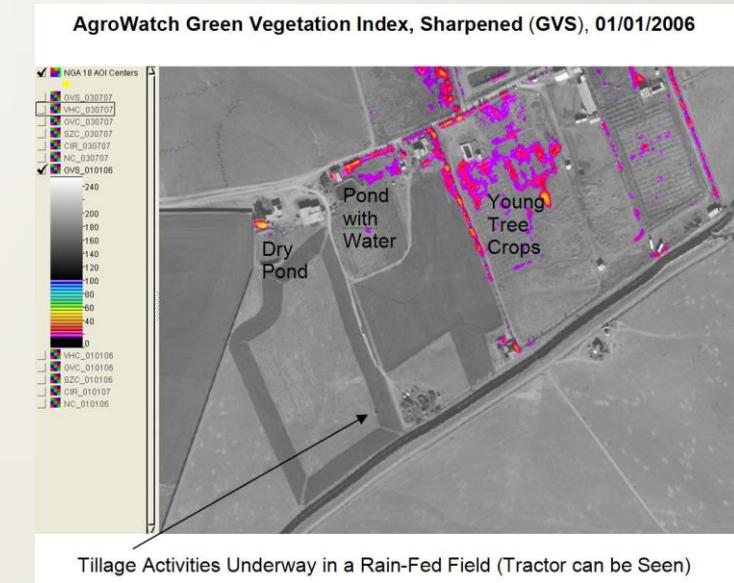
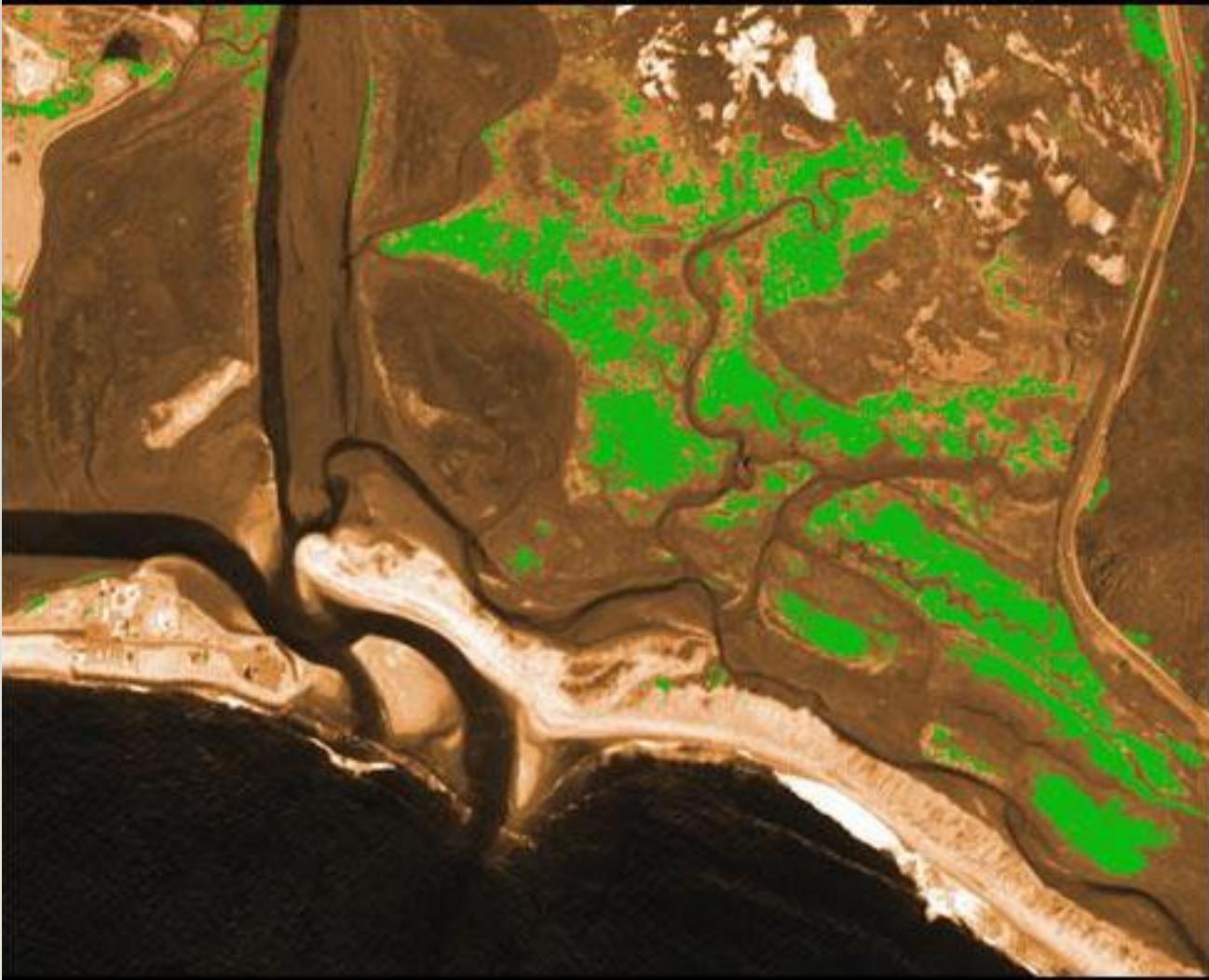
Aerial view of the orchard and its irrigation management units



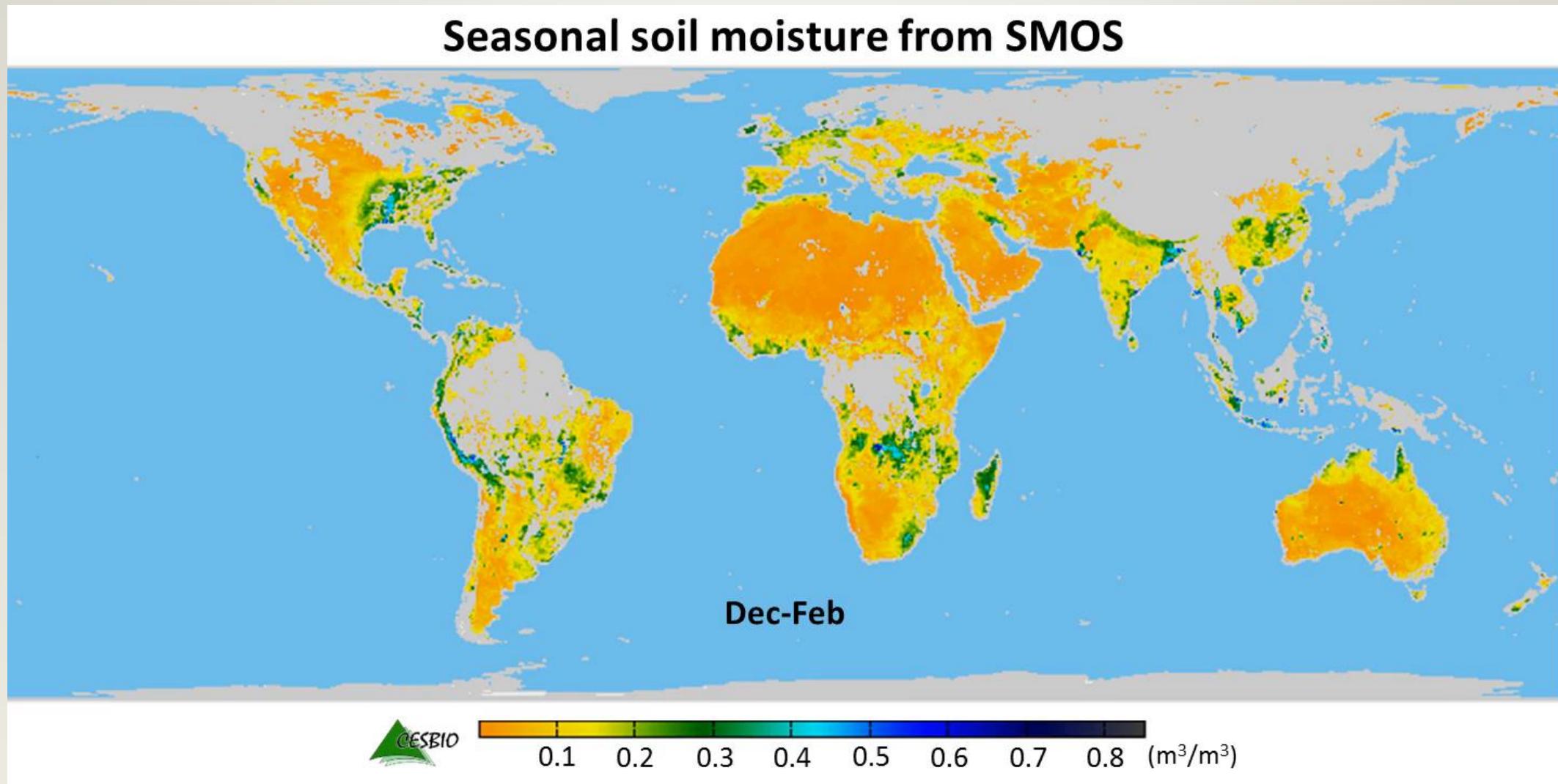
Map of the Crop Water Stress Index (CWSI), showing the scale from minimum to maximum CWSI



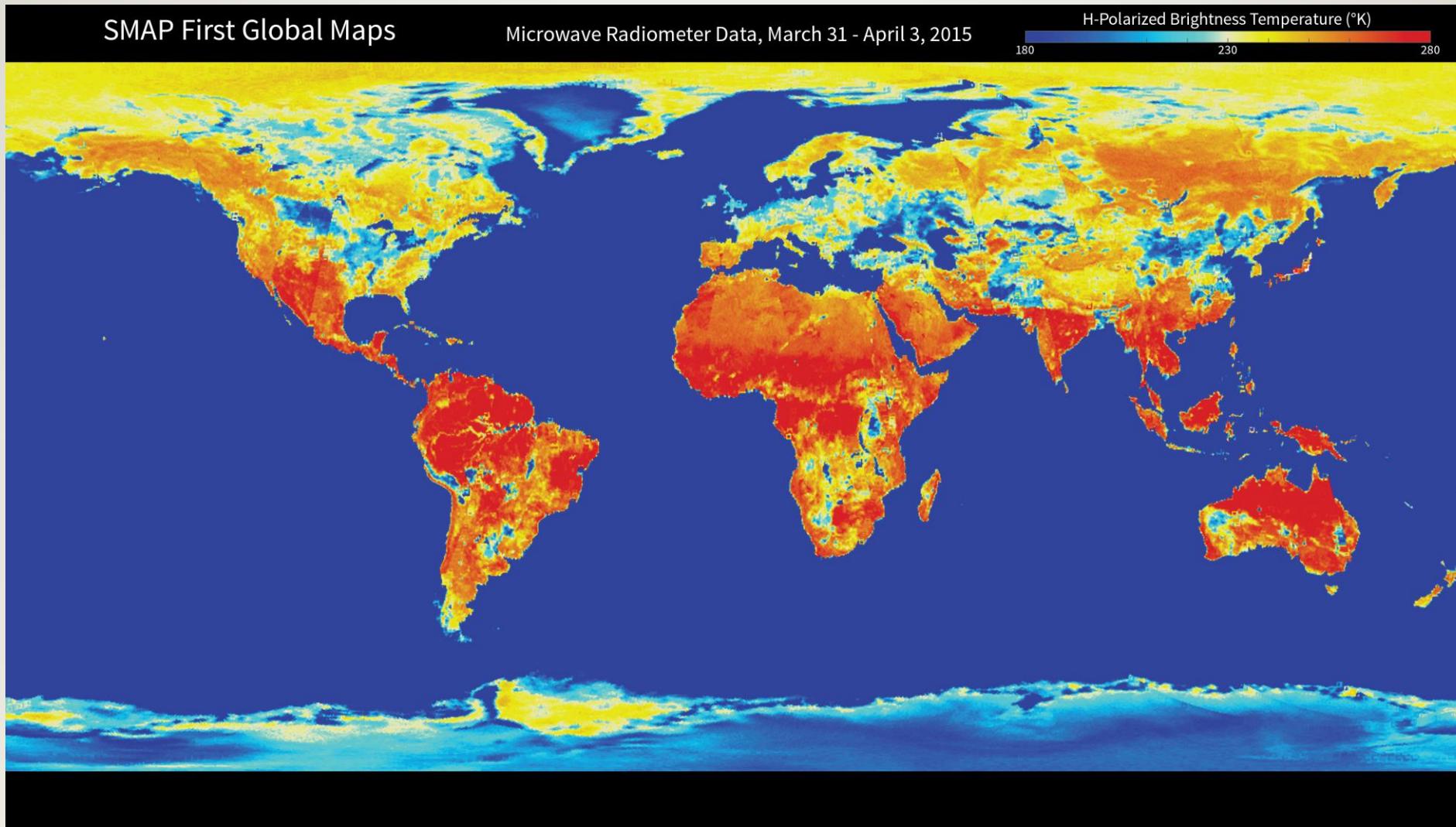
AgroWatch



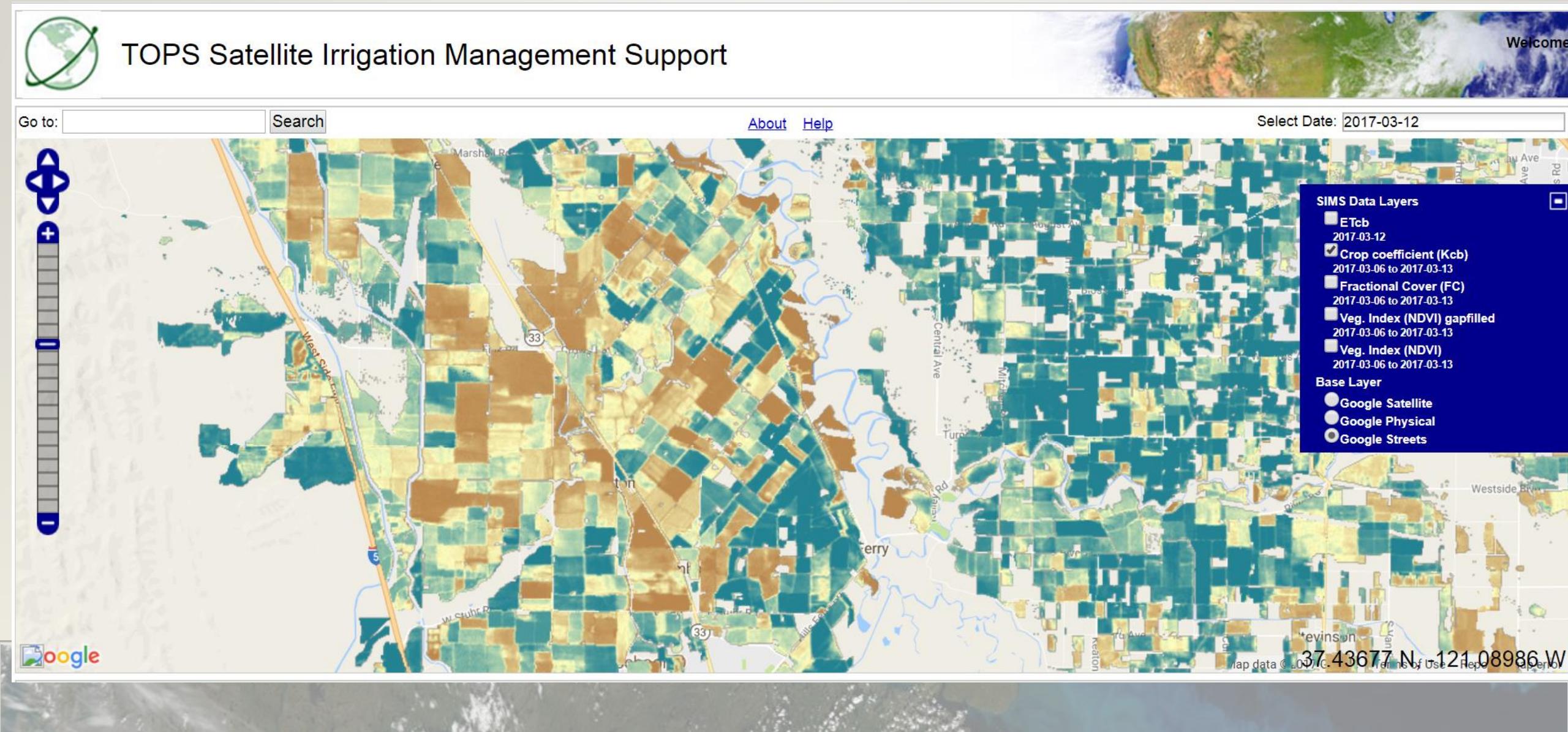
Soil Moisture and Ocean Salinity (SMOS)



Soil Moisture Active Passive (SMAP)



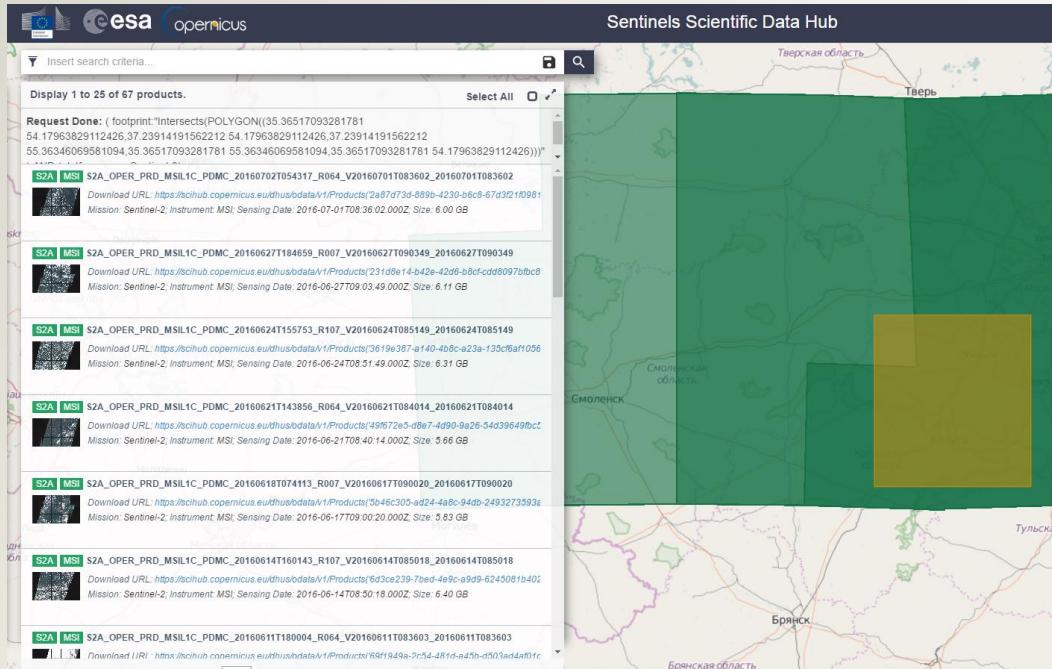
Satellite Irrigation Management Support with the Terrestrial Observation and Prediction System



Water content assessment

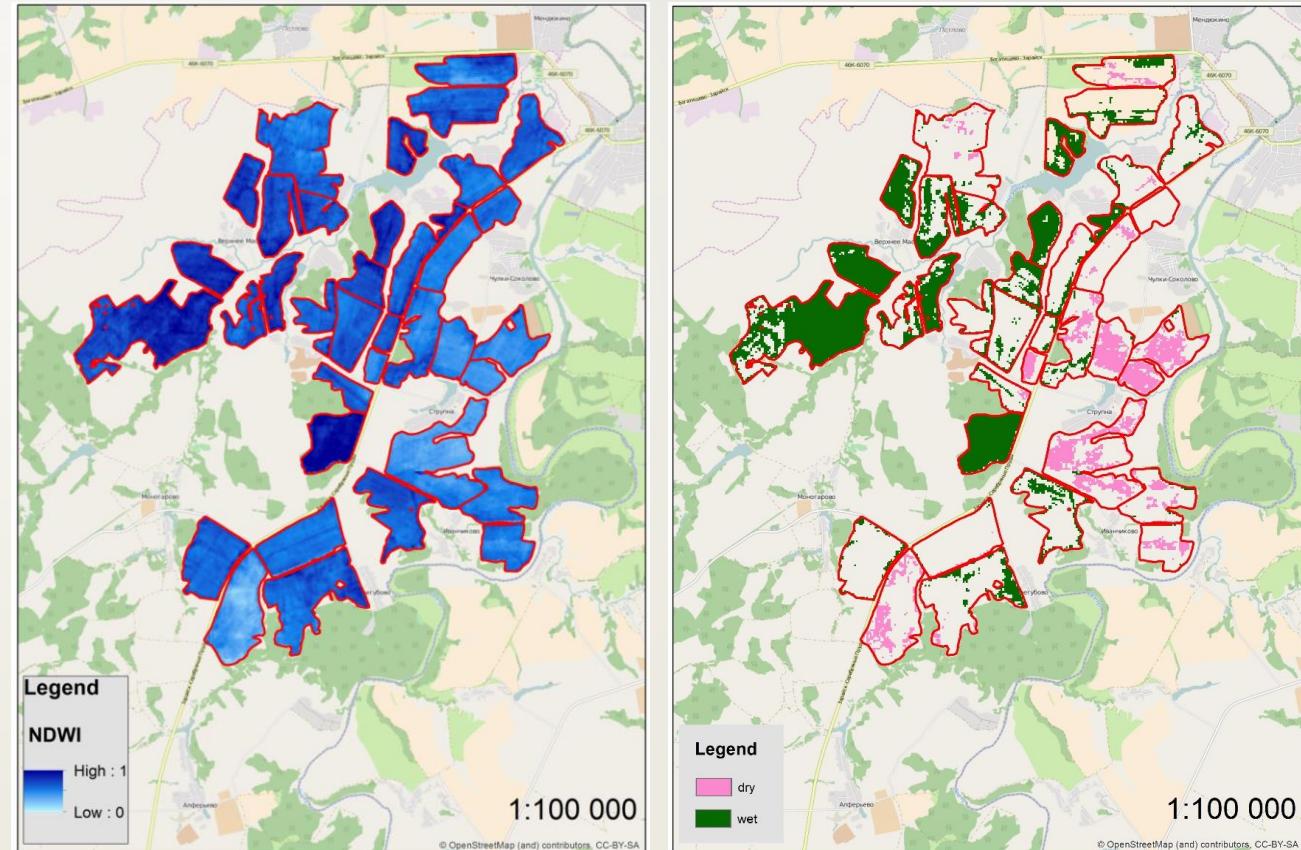
Data: Sentinel – 2

Source: <https://scihub.copernicus.eu/dhus/#/home>
or using QGIS plugin



Based on: The Normalized Difference Water Index (NDWI)

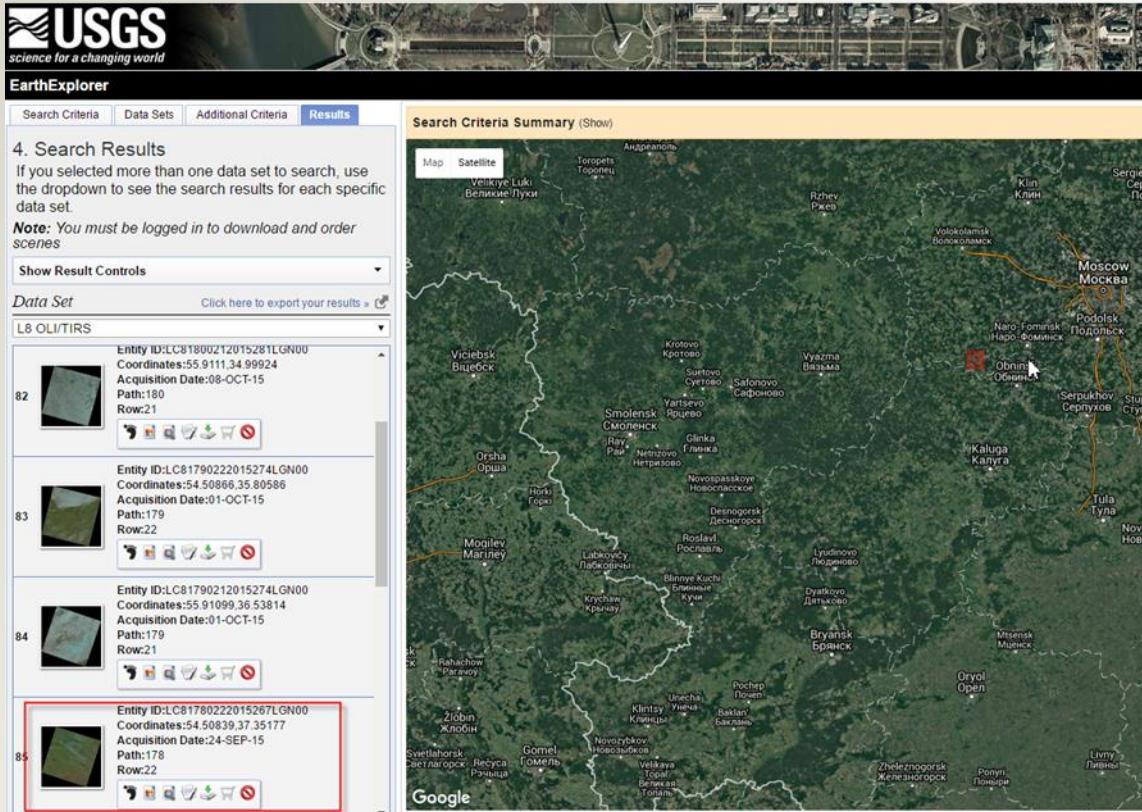
Expression: (NIR - MidIR) / (NIR + MidIR)



Overgrown area detection

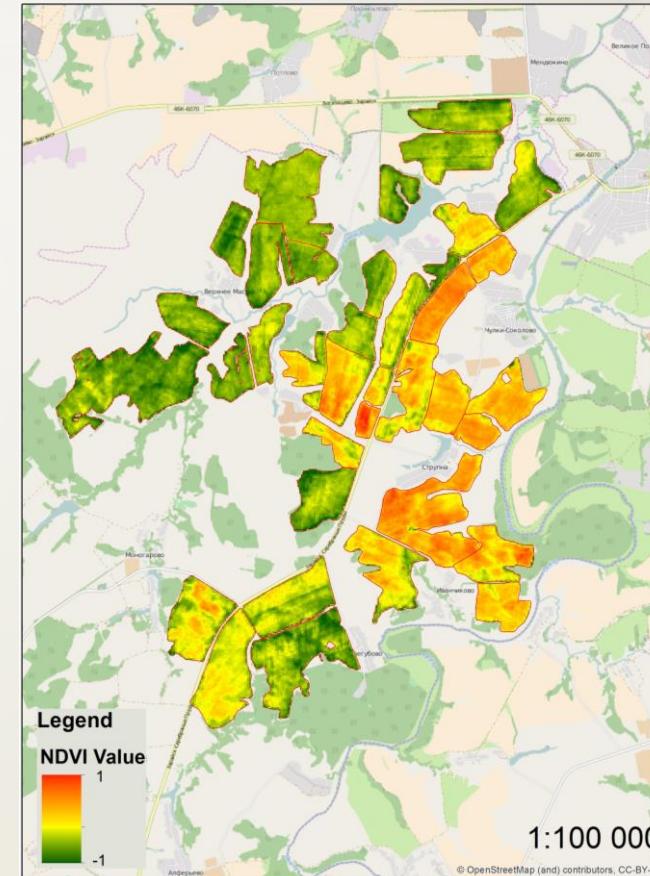
Data: Landsat – 8

Source: <http://earthexplorer.usgs.gov/>



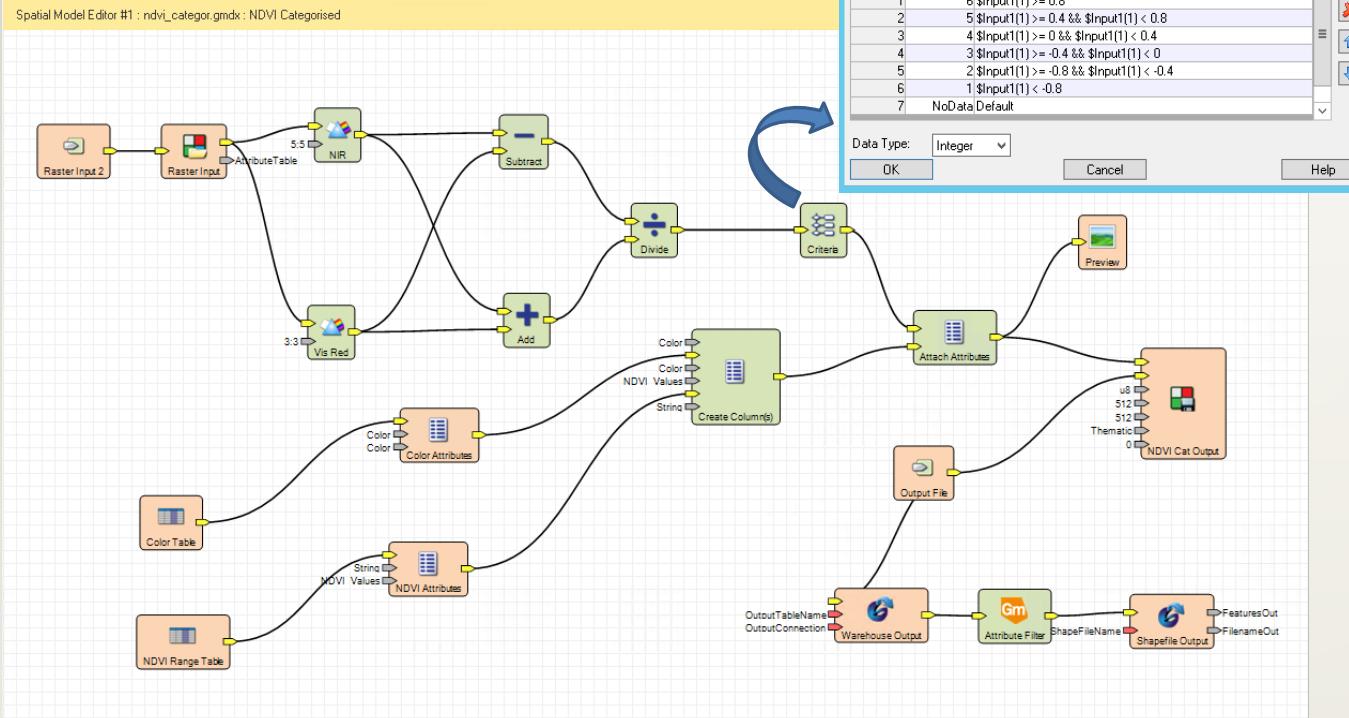
Based on: NDVI index

Expression: $(\text{NIR} - \text{Red}) / (\text{NIR} + \text{Red})$

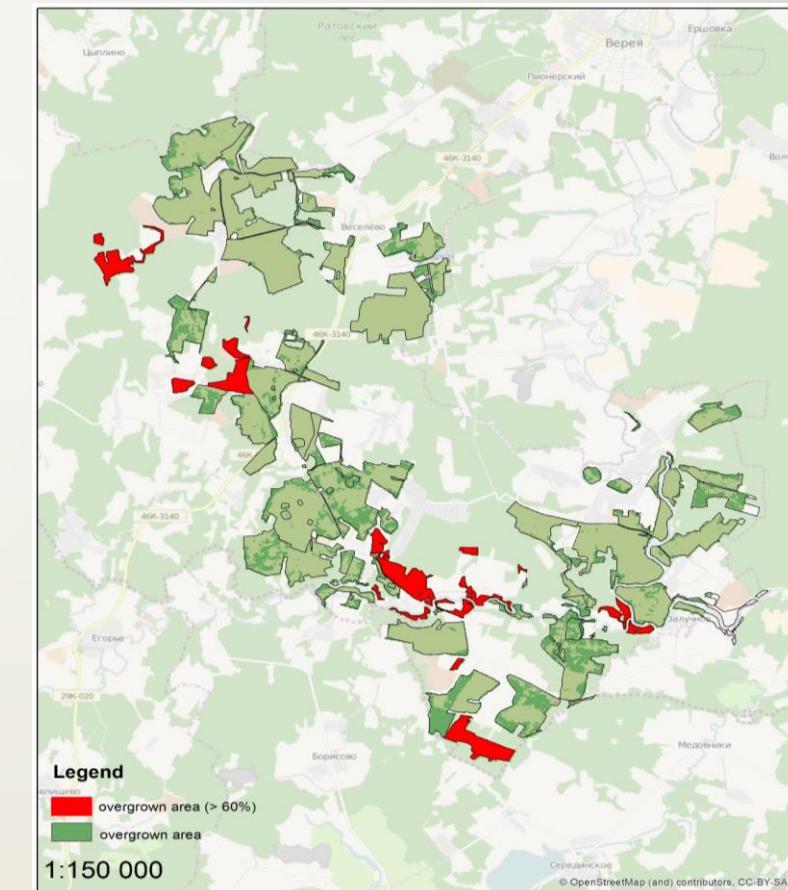


Overgrown area detection

Spatial model for index calculation and overgrown area detection



Overgrown crop parcels



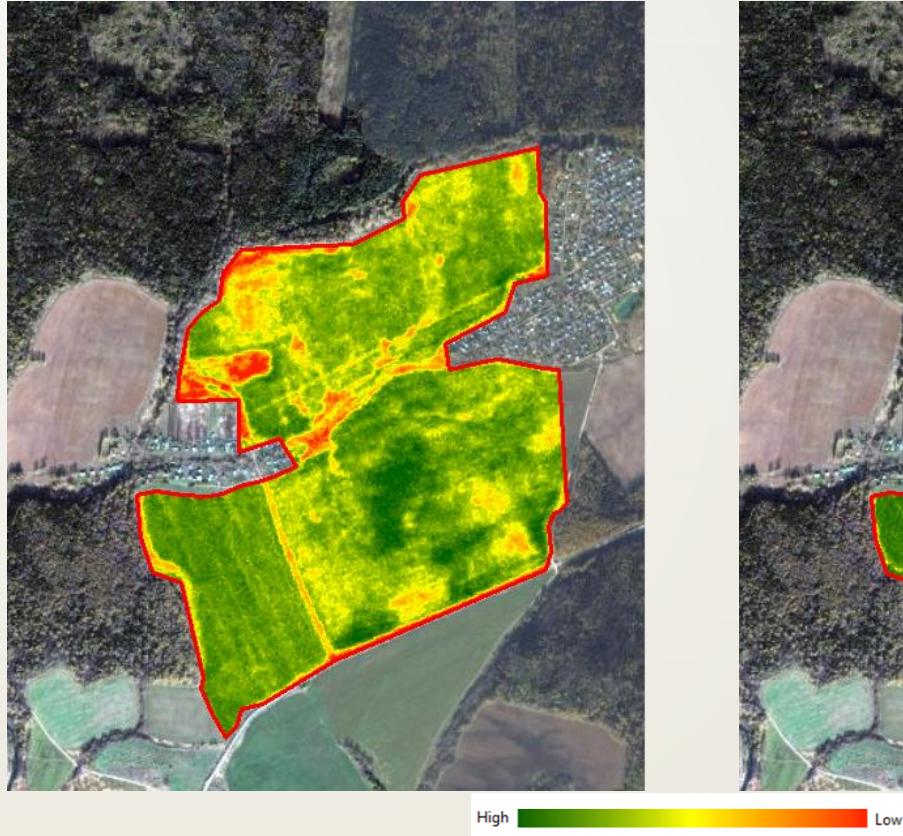
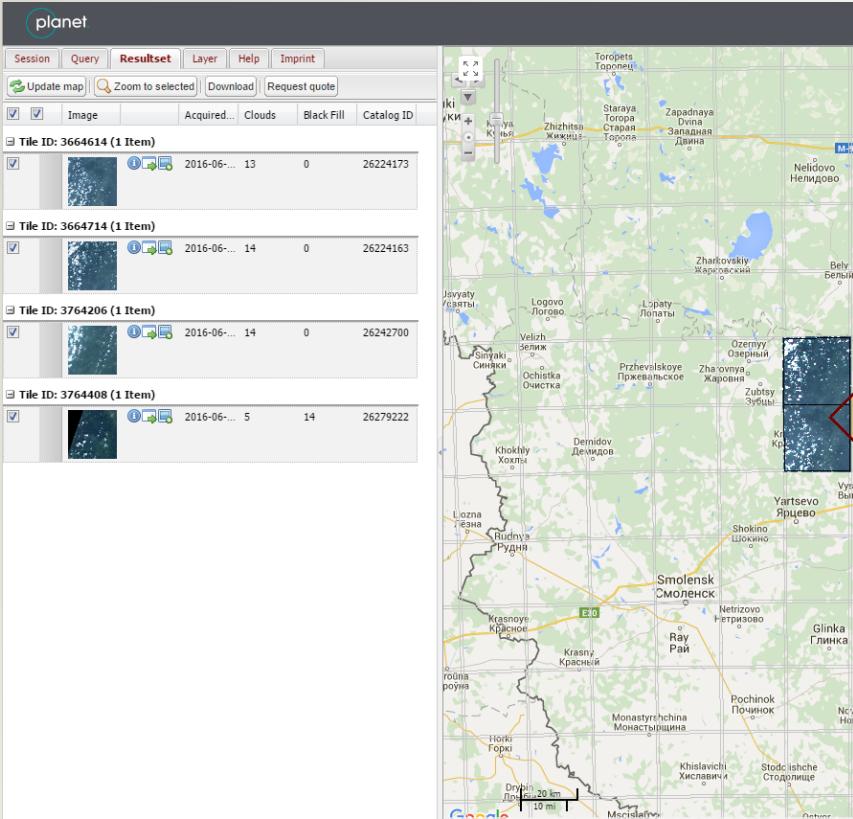
Chlorophyll content assessment

Data: RapidEye

Source: <http://eyefind.rapideye.com/>

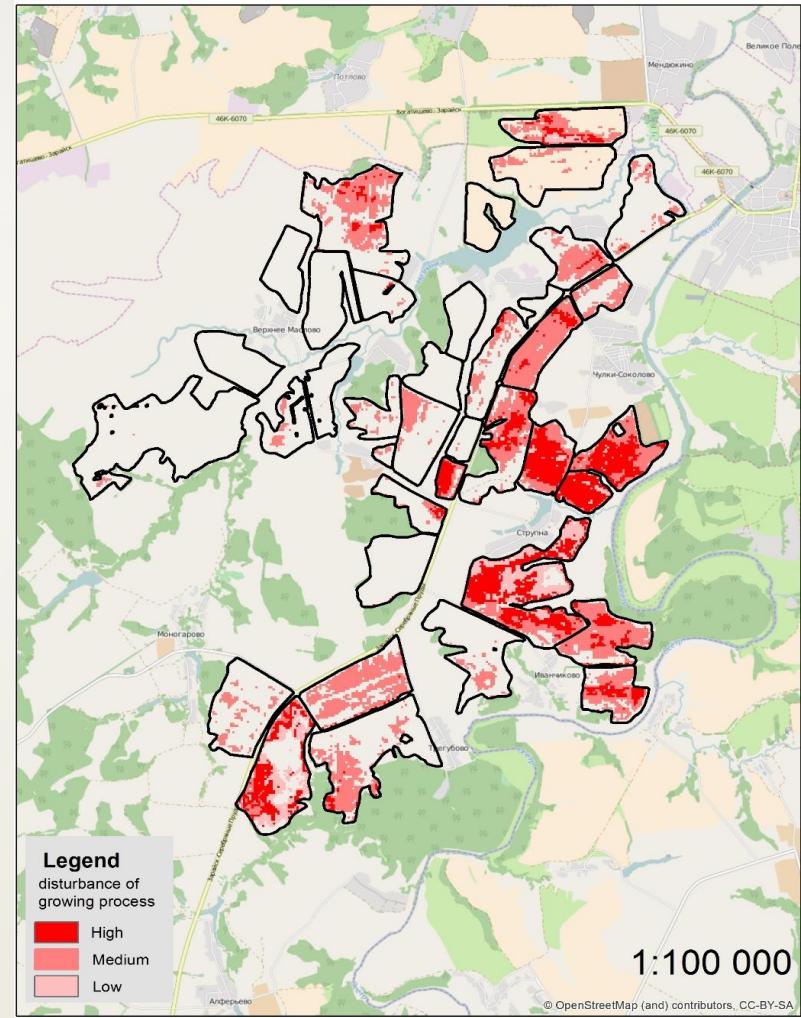
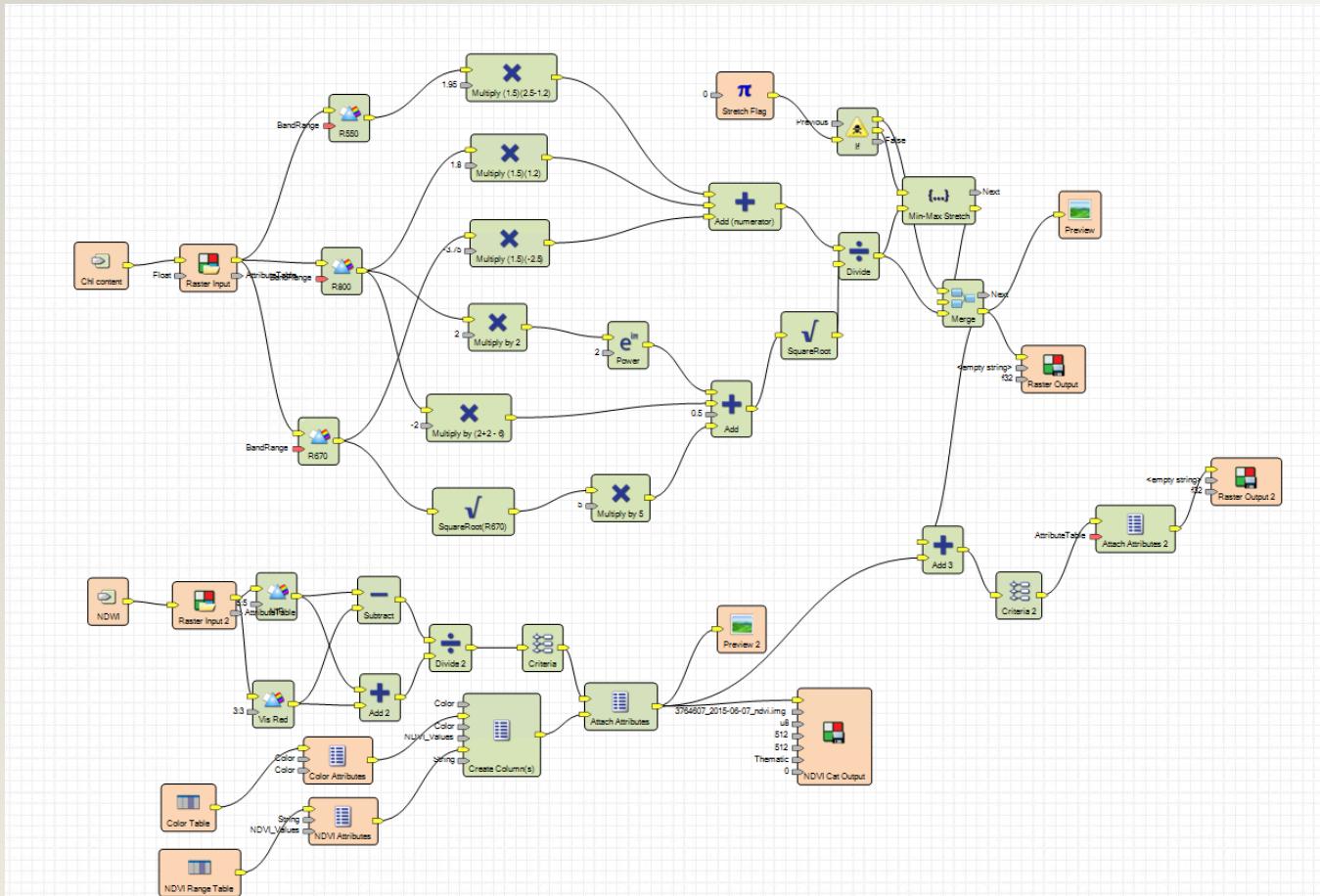
Based on: Improved Modified Chlorophyll Absorption Ratio Index (MCARI)

Expression: $(1.5 * (2.5(NIR - RedEdge) - 1.3(NIR - Green))) / \text{SQRT}((2*NIR + 1)^2 - (6*NIR - 5*\text{SQRT}(RedEdge))) - 0.5$



Relative chlorophyll maps. May and June, 2015

Identification of problematic crop parcels



An aerial photograph of agricultural land featuring a detailed color-coded map overlaid on the fields. The map uses a green-to-red gradient to indicate crop health or yield variability, with darker green areas showing higher health and redder areas showing lower health. The fields are organized into rectangular plots, some of which show distinct patterns of color, suggesting different varieties or stages of growth.

THANK YOU

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